#### Milan, July 30, 2015

### The principles of frequentist and Bayesian medical statistics

Paolo Bruzzi Clinical Epidemiology Unit National Cancer Research Institute Genova - Italy

#### Statistical Mantra

• A study **must** have an adequate size

To warrant an adequate "power" to the study (i.e. to reduce the risk of a false negative result false negative: an effective treatment is not recognised)

To obtain precise estimates of the effects of the experimental therapy

#### **Conventional Statistical Rules**

- A study *must* have an adequate size
- Required Size, based on:
  - Significance level (usually 5%)
  - Power (usually 80-90%)
  - Minimal clinically worthwhile difference

### Sample Size in cancer clinical trials

In trials in early disease, cumulative mortality from 10% to 70%: **500-5000** pts

In trials in advanced disease, cumulative mortality from 50% to 90%: 300-1000 pts

#### **Conventional Statistical Rules**

- A study *must* have an adequate size
- *Required Size: Usually Hundreds/Thousands of patients*
- In many rare cancer conditions: <u>NOT</u>
   <u>POSSIBLE</u>
  - Incidence
  - Age
  - Molecular variant
  - Stage

#### Statistical Mantra

• A study **must** have an adequate size

### **Unjustified Implication**

• If an adequate size cannot be attained, (RARE CANCERS) no methodological ties



### Poor Quality?

- (Study protocol)
- (Classified as Phase II trials)
- No Randomised controls
- Opaque selection of cases
- Primary endpoint: Objective response
- No statistical plan

#### First point to stress

The organization of a trial of **small** size requires **more** care in

- Protocol preparation
- <u>Study design/methodology</u>
- Statistical design
- Addressing Clinical Organizational issues
- ... than a standard size trial

### Methodological issues

- Statistical Power
- Study Design
- Bias in evaluating outcome (double blind)
- Endpoint

#### VALIDITY!

### Study Design

- Uncontrolled trial/Historical Controls
  - Well Kown Biases
  - Sufficient if outstanding benefit
  - Necessary if control group unethical

**Careful and transparent methodology Need of guidelines/research** 

### Study Design

- Uncontrolled trial/Historical controls
- Randomised Controls

WHY NOT?

#### RCT's in rare cancers

#### Pro's

- VALIDITY
- CREDIBILITY

#### Con's

- Moderate loss in power
- Often no standard (untreated control group?)
   Ethics?
  - Acceptance?

#### **Trials in Rare Cancers**

If, despite International cooperation/Prolonged accrual

- it is possible to assemble (in a reasonable time) only <u>a limited number of patients</u>,
- and the efficacy of a new treatment is not outstanding ...

#### What can be done?

#### Recent developments

- Bayesian Statistics
- Surrogate endpoints
- New types of systematic reviews
- Adaptive trials

#### Common beliefs

#### **Frequentist probability**

- Objective
- «Hard»
- Useful to analyse experiments
- Scientific

#### **Bayesian Probability**

- Subjective
- «Soft» mappropriate to analyse experiments
- Not scientific

Differences between Conventional (Frequentist) and Bayesian Statistics

• Meaning of probability

• Use of prior evidence

#### **Frequentist Probability**

Probability <u>of an observation</u> (given a hypothesis)

#### **Bayesian Probability**

Probability **that a hypothesis is true** (given observation and prior knowledge)

#### **Frequentist Probability**

Probability **of the observed difference** (if the experimental therapy does not work)

#### **Bayesian Probability**

Probability <u>that the experimental therapy</u> <u>works/doesn't work (given observed</u> difference and prior knowledge)

#### Frequentist Probability

Definitions and implications

### Probability

Definitions (Wikipedia, from Merriam Webster) ...the measure of the likeliness that an <u>event will</u> occur

- Probability that next number from a roulette will be red = $18/37 \approx 50\%$
- Probability that a man has a cancer in his prostate  $\approx 25/100 = 25\%$
- Probability that my home team (Genoa) had won last game =  $<1/million \approx 0$

- Probability that next number will be red
- Next Number: red = event
- Likeliness = $18/37 \approx 50\%$

Theory:

Red Numbers / Total Numbers = Proportion

Experiment Red Numbers/Plays = **Frequency** 

Probability of an event = proportion

Estimation = frequency = events/plays

- Probability that next number from a roulette will be red = $18/37 \approx 50\%$
- Probability that a man has a cancer in his prostate  $\approx 25/100 = 25\%$ 
  - Proportion: 25 prost.c./100 adult men
  - Estimation: Frequency of examined men in whom I do find a prostate c.:
  - 1 every 4 men (25%)

- Probability that next number from a roulette will be red = $18/37 \approx 50\%$
- Probability that a man has a cancer in his prostate  $\approx 25/100 = 25\%$
- Probability that my home football team (Genoa) had won last game= <1/million ≈ 0</li>
   Proportion? Frequency?

• If my home football team (Genoa) could play again the last game a million times it would not win once

- Theoretical Proportion
- Theoretical Frequency



• Probability that next number from a roulette will be red = $18/37 \approx 50\%$ 

IF (Hypothesis)...

...the roulette works fine (TRUE Frequency = 50%)

- Probability that next number from a roulette will be red = $18/37 \approx 50\%$
- Probability that a man has a cancer in his prostate  $\approx 25/100 = 25\%$
- IF (Hypothesis)...the true prevalence of prostate cancer in Western adult men is 25%

- Probability that next number from a roulette will be red = $18/37 \approx 50\%$
- Probability that man has a cancer in his prostate  $\approx 25/100 = 25\%$
- Probability that my home football team (Genoa) had won last game= <1/million ≈ 0</li>
   IF (Hypothesis)... Genoa does not change his players

#### FREQUENTIST PROBABILITY

The expected frequency

of the observation

given a hypothesis (IF...)

### FREQUENTIST TEST OF HYPOTHESIS

The expected frequency REJECT THE compute of the observation HYPO HESIS Given a hypothesis (IF...) If it is not a If it is too rare event rare

## Other examples of frequentist probabilities

If the roulette works fine (reds = 50%):
Probability that next 3 numbers are red =12%

– Probability that 1/5 numbers are red=19%

- Probability that 1/10 numbers are red = 1%

#### **TESTS of HYPOTHESIS**

Does the roulette work fine (reds = 50%)?
Next 3 numbers are all red P=12% ?

-1/5 numbers are red P=19%?

- 1/10 numbers are red P = 1% (2% if 2-sided) Too rare: I reject the hypothesis that the roulette works fine

# This is what medical statistics is (was) all about!

- 1. Set a hypothesis (null hypothesis,  $H_0$ )
- 2. Do the study
- 3. Compute the probability (frequency, P) of the observed results if  $H_0$  is true
- 4. If p is large, (usually >5%) do not reject
- 5. If p is small (usually < 5 %) reject H<sub>0</sub>

Conventional Statistical Reasoning in Medicine

 Starting Hypothesis = Null Hypothesis, H<sub>0</sub>): <u>New treatment = Standard (no treat.)</u>
 To demonstrate that new treatmt is effective <u>H<sub>0</sub> must be rejected</u>
 To reject H<sub>0</sub>

**Only the results of the trial can be used** 

-> Trials of large size

### **Scientific Method in Medicine** Standard Therapy **New Standard?** Laboratory or trials in other diseases
### **Scientific Method in Medicine Standard Therapy Clinical TRIAL** Laboratory or trials in other diseases

#### **Scientific Method in Medicine** Standard Therapy HO: Exp. No better than standard **Clinical TRIAL** Laboratory or cl. studies



### Scientific Method in Medicine Standard Therapy H0? Laboratory or cl. studies **Clinical TRIAL** H1?New Standard Th.

### Scientific Method in Medicine Standard Therapy H0? Laboratory or cl. studies **Clinical TRIAL** New Standard Th.



#### **Scientific Method in Medicine**

Standard Therapy If Pop P<5% Laboratory Clinical TRIAL or of. studies

New Standard Th.

### **Scientific Method in Medicine Standard Therapy** Po Laboratory **Clinical TRIAL** or cf. studies New Standard Th.

#### **Scientific Method in Medicine Standard Therapy** Laboratory **Clinical TRIAL** or cf. studies New Standa Therapy

#### Advancement of knowledge in Medicine (conventional statistics)

- Dominant theory is true (=standard therapy is better) until sufficient evidence becomes available <u>against</u> it
- To this purpose, only evidence collected within one or more trials aimed at falsifying it can be used
- <u>No use of</u>
  - External evidence
  - Evidence in favor of...

# How to interpret the results of a study

- Internal Validity
- Biological Plausibility
- Internal Coherence
- External Consistency
  - Direct
  - Indirect

Null Hypothesis (H0): the new drug is identical (2) e standard (if no standard, completely ineffective)

- Biological Rationale
- Preclinical studies (disease models)
- Evidence of activity in Phase II
- Evidence of activity within same trial
- Efficacy in other diseases with similar.
- Efficacy in other stages same disease

The 2 Reasons why large numbers of patients are needed in clinical trials

- Outstanding efficacy seldom observed
- <u>Any knowledge outside the primary</u> <u>analysis of the clinical trial is ignored in</u> <u>the design and analysis of the trial</u>

#### **Hypothetical Example**

- As a statistician, I' m asked to design 2 separate trials in the same rare disease, squamous gastric cancer (no standard treat.)
- Study A:

Experimental therapy: Radiochemotherapy

- Effective in squamous cancers of other sites
- Phase II trial; Response Rate 60%
- Study B
- Experimental therapy: Intercessory prayer

#### Squamous gastric cancer

Planning a trial of

RT+CTX

Analysing its results (p value)



### Squamous gastric cancer Planning a trial of me Numbers ame stat RT+C7 Analysing its results (p value)

Squamous gastric cancer Results of the 2 trials RT+CTX Intercessory prayer 20% reduction 20% reduction in deaths =0.15in deaths P=0.15 Treatment x next patient with SGC?

#### Frequentist P

Probability **of the observed difference** if either therapy does not work = 15%

Bayesian Probability Probability <u>that either therapy works a lot/</u> works a little/does not work ? Is it the same for the two treatments? Differences between Conventional and Bayesian Approaches

• Meaning of probability

• Use of prior evidence

#### Conventional P

Probability of the observed difference (<u>if the</u> <u>experimental therapy does not work</u>)

#### **Bayesian Probability**

Probability that the experimental therapy works/doesn't work (given observed difference and **prior knowledge**) Conventional (frequentist) statistical reasoning

Experimental evidence

#### **Bayesian statistical reasoning**

Experimental evidence + Previous Knowledge



# Mortality Tumor X Nil vs A 15% vs 10% N=2000 P = 0.0001

#### H0 Rejected: A is effective in X



Mortality Nil vs A 15% vs 10% Tumor X N=2000P = 0.0001Nil vs A 15% vs 7.5% **Tumor V** N = 240P = 0.066H0 not rejected: A not shown effective in y

Prior Inform X and Y ar	nation: e BRAF+	
		Mortality
Tumor X	Nil vs A	15% vs 10%
N=2000	$\underline{P = 0.0001}$	
Tumor Y	Nil vs A	15% vs 7.5%
N=240	<u>P=0.066</u>	

Prior Inf	ormation:		
X and Y are BRAF+			
A = Anti BRAF		Mortality	
Tumor X	Nil vs A	15% vs 10%	
N=2000	<u>P</u> =	$\underline{P = 0.0001}$	
Tumor Y	Nil vs A	15% vs 7.5%	
N=240	<u>P</u>	<u>P=0.066</u>	
<b>INTERPRETATION?</b>			

#### Interpretation of the two trials

CONVENTIONAL Tumor X: P = 0.0001Tumor Y : P= 0.066Efficacy of treatment A proven in X undemonstrated in Y

#### Interpretation of the two trials

**CONVENTIONAL** Efficacy of treatment A is proven in X, undemonstrated in Y BAYESIAN (Posterior) Probability that treatment A significantly (HR<0.8) lowers mortality in tumor X: 90% in tumor Y: 90%

#### Disadvantages of Bayesian Statistics

- It is (felt as)
  - Subjective
  - Arbitrary
  - Amenable to manipulations (pharma companies?)

#### Conceptual Advantages of Bayesian Statistics

- Reflects human reasoning ("common sense")
- It is focused on estimates of effect
- Provides a conceptual framework for medical decision making
- IT IS TRANSPARENT

#### Practical Advantages of Bayesian Statistics in rare tumors

 No need to set the sample size in advance
 Adaptive designs: enrol patients until sufficient evidence in favour or against efficacy

 When strong a priori evidence is available and trial results are in agreement with it <u>Smaller sample size is necessary – You can</u> <u>stop any time</u>

Note: The difference between Bayesian and conventional statistics decreases with increasing strength of the empirical evidence

**Rare Tumors!** 

• <u>Needed in order to compute posterior</u> <u>probability</u>

- *Needed in order to compute posterior probability*
- It must be transformed into a probability distribution (mean, median, standard deviation, percentiles, etc)

- *Needed in order to compute posterior probability*
- It must be transformed into a probability distribution
- Based on
  - Objective information
  - Subjective beliefs
  - Both

No special way to elicit/obtain prior information

No special way to summarize information - Meta-analytic techniques Frequentist - Bayesian
# Sources of prior evidence

- Randomised Trials
- Biological & Preclinical Studies
- Case-reports
- Uncontrolled studies
- Studies with surrogate endpoints
- Studies on other similar cancers
- Studies on the same cancer in different stages
- Others?

## Meta-analyses in frequent tumors

- Randomised Trials
- Biological & Preclinical Studies
- Case-reports
- Uncontrolled studies
- Studies with surrogate endpoints
- Studies on other similar cancers
- Studies on the same cancer in different stages
- Others?

## Meta-analyses in frequent tumors

### - Randomised Trials

Weighted exclusively based on their size (and quality)

## Rare Tumors

- Kanaomiseu Trials
- Biological & Preclinical Studies
- Case-reports
- Uncontrolled studies
- Studies with surrogate endpoints
- Studies on other similar cancers
- Studies on the same cancer in different stages
- Others?

## Meta-analyses in rare tumors

Need to use information from studies

- <100% valid
- <100% pertinent to the question of interest
  - Different cancers
  - Different treatments
  - Different endpoints

## Prior evidence and clinical trials

Need to develop and validate new (metaanalytic) approaches to summarize prior information in rare tumors

#### Requirements

- Explicit
- Quantitative
- Reproducible

## Efficacy trials in rare tumors

- Uncontrolled efficacy (phase III) trials of high quality
- Randomized activity (Phase II) trials followed by uncontrolled efficacy trials (with historical controls)
- -RCT's with surrogate endpoints
- <u>Adaptive, Bayesian, activity/efficacy</u>
   <u>RCT's based on unconventional</u>
   <u>Systematic Reviews</u>

## RCT's in rare cancers

Loss of power (50% less patients in exp treatment)
Available patients : 100
Cure Rate in controls: 40%

RCT (50 x2):80% power for delta:30% (to 70%) Uncontr.trial 80% power for delta:21% (to 61%)

## RCT's in rare cancers

Loss of power /Precision
(50% less patients in exp treatment)
Available patients : 100

RCT (50 x2): Uncontrolled tr. (Histor. Controls)

Difference +/- 15% Difference +/- 11% Differences between the present and the proposed approach

- Present :
  - Rational but informal integration of the available knowledge
- Proposed (Bayesian)
  - Formal, explicit and quantitative integration of the available knowledge
    - Verifiable quantitative methods
    - Sensitivity analyses
    - Focus on summary effect estimates